PI Controller Design Using Model Reference Adaptive Control Approaches
For A Chemical Process

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Abstract—This paper discusses the application of Model Reference Adaptive Control (MRAC) concepts in designing an adaptive feedback controller to tune a given PI controller. The approaches of using the gradient (MIT) and stability (Lyapunov) methods are shown. The effectiveness of the two methods are shown through simulation and comparison is made to show which method give the best result. The results show that the stability method produces a better result.

Keywords — Model Reference Adaptive Control; gradient approach, MIT rule; stability approach, Lyapunov method

I. INTRODUCTION

In every day language “to adapt” means to change a behaviour to conform to new circumstances. At an early symposium in 1961 a long discussion ended with the following suggestion, an adaptive system is any physical system that has been designed with an adaptive viewpoint [1]. In this paper we deals with the pragmatic attitude to design a PI controller with adjustable parameters and a mechanism for adjusting the parameters by using the gradient approach (MIT) and the stability approach.

The aim of this assignment is to design two controllers for a chemical process based on the gradient and stability approaches. The performance of both will be compared via computer simulation to show the effectiveness of the controllers.

The paper is organized as follows. The MRAC theory of gradient method is presented in section III A and the stability approach is presented in section III B. The adaptive feedback controller scheme is presented in section IV. The simulation results are shown in section V. Finally, the results are concluded at the end of the paper.

II. PI CONTROLLER

An adaptive Proportional-Integral (PI) controller is required to control a chemical process. In PI-controller, there is another term in the controller equation [6].

\[ u(t) = K \left( e(t) + \frac{1}{T_i} \int_0^t e(t) dt \right) \]

where \( T_i \) is integration time constant. If the controller is tuned to be slow and \( T_i \) is large, then the controller first acts like P-controller, but later the as the integration starts to take effect, the steady-state deviation goes slowly to zero. In PI-control, the steady-state deviation will finally go to zero. If the controller is tuned to be fast and \( T_i \) is small, then both terms (P and I) will affect the control signal all the way from the beginning. The system becomes faster, but the output signal might oscillates.

III. MODEL REFERENCE ADAPTIVE CONTROL

Fig. 1 illustrates the general structure of the Model Reference Adaptive Control (MRAC) system. The basic MRAC system consists of 4 main components:

i) Plant to be controlled
ii) Reference model to generate desired closed loop output response
iii) Controller that is time-varying and whose coefficients are adjusted by adaptive mechanism
iv) Adaptive mechanism that uses ‘error’ (the difference between the plant and the desired model output) to produce controller coefficient

Regardless of the actual process parameters, adaptation in MRAC takes the form of adjustment of some or all of the controller coefficients so as to force the response of the resulting closed-loop control system to that of the reference model. Therefore, the actual parameter values of the controlled system do not really matter.

A. The Gradient Approach

The Gradient Approach/Method of designing an MRAC controller is also known as the MIT Rule as it was first
developed at the Massachusetts Institute of Technology (MIT), USA. This is the original method developed for adaptive control design before other methods were introduced to overcome some of its weaknesses. However, the Gradient method is relatively simple and easy to use.

In designing the MRAC controller, we would like the output of the closed-loop system \( y \) to follow the output of the reference model \( y_m \). Therefore, we aim to minimise the error \( e = y - y_m \) by designing a controller that has one or more adjustable parameters such that a certain cost function is minimised.

B. The Stability Approach

MRAC can be designed such that the globally asymptotic stability of the equilibrium point of the error difference equation is guaranteed. To do this, the Lyapunov Second Method is used. It requires an appropriate Lyapunov function to be chosen, which could be difficult. This approach has stability consideration in mind and is also known as the Lyapunov Method.

IV. ADAPTIVE FEEDBACK CONTROLLER DESIGN

The chemical process is described by the transfer functions and a controller by the form as indicated below where the time delay \( T_D \) and the time constant \( \tau \), of the process are unknown.

**Plant:**

\[
G(s) = e^{-T_ds} \frac{1}{ts+1}
\]

**Reference model:**

\[
G_m(s) = e^{-s} \frac{1}{5s+1}
\]

**Controller:**

\[
u = -ky + k_\tau (r - y)
\]

A. Time Delay Approximation

A time delay element is represented by a non-linear transfer function. Since many control design methods require linear systems, it is often necessary to approximate a time delay by a linear transfer function \[3,6\]. Therefore, a suitable approximation is needed to represent the time delay term in the system.

There are many approximation methods available but for this assignment, we have selected Padé approximation. The Padé approximation for the time delay is given below:

\[
e^{-T_ds} = \frac{-T_ds + 2}{T_ds + 2}
\]

\[
e^{-\tau} = \frac{-s + 2}{s + 2}
\]

The above approximations will be used throughout this assignment in designing the controller.

B. The Gradient Approach Design

A closed-loop system with a controller that has the following parameters:

\[
\begin{align*}
 r(t) &= \text{Reference input signal} \\
u(t) &= \text{Control signal} \\
y(t) &= \text{Plant output} \\
y_m(t) &= \text{Reference model output} \\
e(t) &= \text{Difference between plant and reference model output} = y(t) - y_m(t)
\end{align*}
\]

The control objective is to adjust the controller parameters, \( \theta_1 \) and \( \theta_2 \), so that \( e(t) \) is minimised. To do this, a cost function, \( J(\theta) \) is chosen and minimised. Apart from Eqns. (1) through (5), the cost function chosen is of the form

\[
J(\theta) = \frac{1}{2} e^2
\]

Let \( \theta_1 = \frac{k_1}{s} \) and \( \theta_2 = \frac{s k + k_1}{s} \). Replacing them into Eqn. (3), thus the controller, \( u \) becomes

\[
u = \theta_1 r - \theta_2 y
\]

After time delay approximation using Eqns. (4) and (5), thus Eqns. (1) and (2), each becomes

**Plant:**

\[
G(s) = \frac{-T_ds + 2}{\tau T_ds^2 + (2\tau + T_D)s + 2 + (-T_Ds + 2)\theta_2}
\]

**Reference model:**

\[
G_m(s) = \frac{-s + 2}{5s^2 + 11s + 2}
\]

Notice that from Eqn. (8),

\[
y = \frac{-T_ds + 2}{\tau T_ds^2 + (2\tau + T_D)s + 2 + (-T_Ds + 2)\theta_2} u
\]

Then, after substituting Eqn. (7) into Eqns. (10),

\[
y_m = \frac{-s + 2}{5s^2 + 11s + 2}
\]

Therefore the error, \( e = y - y_m \)

\[
\begin{align*}
e &= \frac{(-T_ds + 2)\theta_2 r}{\tau T_ds^2 + (2\tau + T_D)s + 2 + (-T_Ds + 2)\theta_2} \left( -s + 2 \right)^2 \\
&= \frac{\left( -T_ds + 2 \right) \theta_2 r}{s^2 + 11s + 2}
\end{align*}
\]

So we have,

\[
\frac{d\theta_1}{dt} = -\gamma \cdot e \cdot \frac{(-T_ds + 2)r}{\tau T_ds^2 + (2\tau + T_D)s + 2 + (-T_Ds + 2)\theta_2}
\]

and

\[
\frac{d\theta_2}{dt} = \gamma \cdot e \cdot \frac{(-T_ds + 2)r}{\tau T_ds^2 + (2\tau + T_D)s + 2 + (-T_Ds + 2)\theta_2}
\]

In this case we need to do some approximation: i.e. perfect model following, \( y = y_m \). Therefore, we then have,
\[
\frac{d\theta_1}{dt} = -\gamma \cdot e \cdot r \left[ \frac{-T_D s + 2}{5s^2 + 11s + 2} \right]
\]
\[
\theta_1 = \frac{-\gamma \cdot e \cdot r}{s} \left[ \frac{-T_D s + 2}{5s^2 + 11s + 2} \right] = k_i
\]
\[
\frac{d\theta_2}{dt} = \gamma \cdot e \cdot r \left[ \frac{-T_D s + 2}{5s^2 + 11s + 2} \right] = k_i/s
\]
\[
\theta_2 = \frac{\gamma \cdot e \cdot r}{s} \left[ \frac{-T_D s + 2}{5s^2 + 11s + 2} \right] = k_i/s
\]

Which results in,
\[
k = \gamma \cdot e \cdot (r + y) \left[ \frac{-T_D s + 2}{5s^2 + 11s + 2} \right]
\]
\[
k_i = \gamma \cdot e \cdot r \left[ \frac{-T_D s + 2}{5s^2 + 11s + 2} \right]
\]

The MRAC gradient approach is then simulated using Matlab SIMULINK block with the reference input of a square wave signal with amplitude 3 and frequency 0.0033Hz. Both the output of the system responses \( (y \text{ and } y_m) \) are coupled and placed onto scope for simultaneous viewing. The scope is used to simultaneously observe the output response of \( y \) and the desired output response of \( y_m \).

Fig. 2 below shows the system output responses for a MIT rule based design with adaptation gains, \( \text{gain1}=0.02 \text{ and gain2}=0.0015 \).

C. The Stability Approach Design

In designing an MRAC using Lyapunov Method, the following steps should be followed:

i) Derive a differential equation for error, \( e = y - y_m \) (i.e. \( \dot{e}, \ddot{e} \text{ etc.} \) that contains the adjustable parameter, \( \theta \).

From Eqn. (11), after replacing \( A = -T_D \), \( B = \tau T_D \) \( C = 2\tau + T_D \) and \( d = \frac{1}{B} \), the differential equation becomes
\[
\ddot{y} = -(A\theta_2 + C)\dot{y} + (2 + 2\theta_2)ky + A\theta_1 dr + 20r,dr
\]
\[
\dddot{y}_m = -2.2\dot{y}_m - 0.4y_m - 0.2r + 0.4r
\]

Substituting Eqns. (14) and (15) into \( \ddot{e} = \dddot{y} - \dddot{y}_m \), thus
\[
\ddot{e} = -2.2\dot{e} - 0.4e - \left( \theta + \frac{C}{A} \frac{2}{2d} A \ddot{y} - \left( \theta + \frac{10}{2} \right) d \right) ky
\]
\[
+ \left( \frac{2}{Ad} + \theta_1 \right) Adr + \left( \theta - 0.4 \right) \frac{2}{2d} dr
\]
(16)

ii) Find a suitable Lyapunov function, \( V(e,\theta) \) - usually in a quadratic form (to ensure positive definiteness).

The Lyapunov function, \( V(e,\theta) \), is based on (16). \( V = \lambda_1 e^2 + \lambda_2 e^2 + \lambda_3 X_1^2 + \lambda_4 X_1^2 + \lambda_5 X_2^2 + \lambda_6 X_2^2 \), where \( \lambda_1, \lambda_2, \lambda_3, \lambda_4, \lambda_5, \lambda_6 > 0 \) is positive definite. The derivative of \( V \) becomes
\[
\dot{V} = -4.4\lambda_1 \dot{e}^2 + (\lambda_2 - 0.4\lambda_1) \dot{e}e + (\lambda_3 \dot{X}_1 - \lambda_3 \dot{e}X_1 + (\lambda_4 \dot{X}_1 - 2\lambda_3 \dot{e}y)X_1,
\]
\[
+ (\lambda_5 \dot{X}_2 + \lambda_5 \dot{e} \dot{X}_2) X_2 + (\lambda_6 \dot{X}_2 + 2\lambda_5 \dot{e} \dot{X}_2) X_1 + (\lambda_6 \dot{e} \dot{X}_2 + 2\lambda_5 \dot{e} \dot{X}_2) X_1,
\]
where for stability \( V \) must be positive i.e. \( \dot{V} < 0 \).

iii) Derive an adaptation mechanism based on \( V(e,\theta) \) such that \( e \) goes to zero.

\[
\dot{X}_1 = \frac{2.5\lambda_2 \dot{Ad} \dot{e}}{\lambda_3} \left( \frac{1}{\lambda_3} \right)
\]
\[
\dot{X}_2 = \frac{5\lambda_6 \dot{e} \dot{X}_1}{\lambda_4} \left( \frac{1}{\lambda_4} \right)
\]
\[
\dot{X}_3 = -\frac{2.5\lambda_2 \dot{Ad} \dot{e}}{\lambda_5} \left( \frac{1}{\lambda_5} \right)
\]
\[
\dot{X}_4 = -\frac{5\lambda_6 \dot{e} \dot{X}_1}{\lambda_6} \left( \frac{1}{\lambda_6} \right)
\]

Summing Eqns. (17) and (18), results in
\[
\dot{\theta}_2 = 1.25 \dot{e} \left( \frac{2 \lambda_3 y}{\lambda_3} + \frac{2 \lambda_3 y}{\lambda_4} \right)
\]
\[
\dot{\theta}_2 = k + \frac{k_i}{s} \left( \frac{\dot{e} \dot{y} + \frac{2 \lambda_3 y}{\lambda_4}}{\frac{2 \lambda_3 y}{\lambda_4}} \right)
\]

Similarly, summing Eqns. (19) and (20) results in
\[
\dot{\theta}_1 = -2.5 \lambda_2 \lambda_6 \dot{Ad} \dot{e} + 5 \lambda_2 \lambda_5 \dot{e} \dot{d} \left( \frac{1}{2 \lambda_5 \lambda_6} \right)
\]
\[
\dot{\theta}_1 = k + \frac{k_i}{s} \left( \frac{\dot{e} \dot{r} + \frac{2 \lambda_5 y}{\lambda_6}}{\frac{2 \lambda_5 y}{\lambda_6}} \right)
\]

Therefore,
\[
k = -\gamma_1 \dot{e}(\dot{y} - \dot{y}_m) - \gamma_2 \dot{e}(\dot{r} - 4r)
\]
\[
k_i = \gamma_2 \dot{e}(\dot{r} - 4r)
\]
Fig. 3 below shows the outputs for $y$ and $y_m$ for MRAC Lyapunov Method based design with the adaptation gain value, $\gamma_1$ & $\gamma_2$ with 0.3 and 0.1 respectively. The output response of the system is displayed simultaneously in the scope similar to Fig. 1. The outputs for $y$ and $y_m$ are stored in array in MATLAB workspace for further analyses.

V. RESULTS COMPARISON AND DISCUSSION

The MRAC designed using the Gradient method/MIT rule has been shown that it does not guarantee stability to the resulting closed-loop system. On the other hand, designing an MRAC using the stability approach will ensure a stable closed loop system. Based on observation from the above designs, the differences of each design can be highlighted as below:

1. The adjustable parameter $\theta_1$ and $\theta_2$ for Lyapunov method are simpler than that of the MIT rule. The parameters of each adjustable design are summarized in Table 1.

<table>
<thead>
<tr>
<th>TABLE I. ADJUSTABLE PARAMETER FOR MRAC DESIGN</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lyapunov Method</td>
</tr>
<tr>
<td>[ \frac{d\theta_1}{dt} = -\gamma_1 er ]</td>
</tr>
<tr>
<td>[ \frac{d\theta_2}{dt} = \gamma_2 e y ]</td>
</tr>
</tbody>
</table>

2. The performance of MRAC designs depend on the amplitude of the reference input signal, where the increase of amplitude from 1 to 3 will effect the above designs. It can be seen from Fig. 4 (a) and (b), that although the controllers designed based on both Gradient and Lyapunov Methods perform very well with r(t) of amplitude 1, when the amplitude of r(t) is changed to 3, the Gradient Method does not give very satisfactory results as the overshoots during transient is very high. This could result in the control actuator having to work extra hard at every cycle.

VI. CONCLUSIONS

In conclusion, designing an adaptive controller through stability approach can ensure a stable closed loop system whereby the gradient approach does not take any consideration on the stability of the system. The stability approach produces a better result.

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